

# Convergence of random products of Markov transition functions

By S. R. FOGUEL in Jerusalem (Israel)

## 1. Introduction

The purpose of this note is to apply a theorem of AMEMIYA and ANDO [1] to study a Markov process  $\{x_n\}$  with the phase space  $(X, \Sigma)$  which is not stationary but *there exists a mapping  $n \rightarrow r(n)$  of the set of all non-negative integers onto the finite set  $\{1, \dots, N\}$  such that the conditional probabilities*

$$\Pr \{x_{n+1} \in A | x_n = x\} \quad (x \in X, A \in \Sigma, n = 0, 1, 2, \dots)$$

*depend (besides on  $A$  and  $X$ ) only on  $r(n)$ , i.e.*

$$\Pr \{x_{n+1} \in A | x_n = x\} = P_{r(n)}(x, A).$$

By means of  $P_j(x, A)$  ( $j = 1, \dots, N$ ) one defines in the Banach space of the finite measures on  $(X, \Sigma)$  operators  $\nu \rightarrow \nu P_j$  by

$$(1.1) \quad (\nu P_j)(A) = \int P_j(x, A) \nu(dx).$$

Thus if  $\mu$  is a probability measure on  $(X, \Sigma)$  and we choose it to be the distribution of  $x_0$  then the distribution  $\mu_n$  of  $x_n$  is, as well known,

$$\mu_n = \mu P_{r(0)} \dots P_{r(n-1)}.$$

We shall prove the convergence of  $\mu_n(A)$  as  $n \rightarrow \infty$  under suitable conditions.

## 2. Convergence of random products

Let the mapping  $n \rightarrow r(n)$  be defined on the set of all non-negative integers and assume the values  $1, 2, \dots, N$ . Let  $P_j(x, A)$  be subtransition functions on the measure space  $(X, \Sigma, \lambda)$  with  $\lambda$  being a  $\sigma$ -finite subinvariant measure:  $P_j(x, A)$  is a function on  $X \times \Sigma$  which is for each  $x \in X$  a non-negative measure of total measure  $\leq 1$  and, for each  $A \in \Sigma$ , a measurable function and

$$(2.1) \quad \int P_j(x, A) \lambda(dx) \leq \lambda(A).$$

It is well known that  $P_j(x, A)$  induce contraction operators on  $L_2(X, \Sigma, \lambda)$  by

$$(2.2) \quad (P_j f)(x) = \int P_j(x, dy) f(y)$$

From now on  $P_j$  will be considered as contraction operators on  $L_2$  only, also every relation between functions will be a. e.

In [1] it is proved that if each  $P_j$  satisfies the condition:

$$(W') \quad \|P_j f\| = \|f\| \text{ implies } P_j f = f$$

and each  $1 \leq j \leq N$  appears infinitely often in the sequence  $r(n)$  then the sequences of operators  $P_{r(n)} \dots P_{r(1)}, P_{r(1)} \dots P_{r(n)}$  both converge weakly to the projection on the intersection  $\bigcap_{j=1}^N \{f: P_j f = f\}$ . Let us study property (W') in our case.

**Lemma 1.** Let  $P(x, A)$  be a subtransition function on  $(X, \Sigma, \lambda)$  with  $\lambda$  being a subinvariant measure. Let  $K = \{f: \|Pf\| = \|f\|\}$ . Then  $K$  is generated by characteristic functions of sets of finite measure and if  $1_A \in K$  for some  $A \in \Sigma$  then  $P(x, A)$  assumes the values zero and one only.

**Proof.** The equation  $\|Pf\| = \|f\|$  is equivalent to  $P^*Pf = f$  since  $P$  is a contraction operator. Thus  $K$  is a subspace of  $L_2$ .

Now if  $f \geq 0$  then  $Pf \geq 0$ . Also if  $f \geq 0$  then  $P^*f \geq 0$ : otherwise, if  $P^*f < 0$  on a set  $A$  of positive finite measure, then  $0 > \int_A P^*f d\lambda = \int P 1_A \cdot f d\lambda \geq 0$ . Thus if  $f \in K$  and is real,  $P^*P|f| \geq |P^*Pf| = |f|$ . Inequality is impossible since  $\|P^*P\| \leq 1$ , hence  $|f| \in K$ . Now if  $0 \leq f \leq c$  then  $0 \leq Pf \leq c$ , also  $P^*f \leq c$ : otherwise, if  $P^*f > c$  on a set of positive finite measure  $A$ , then

$$c\lambda(A) < \int_A P^*f d\lambda = \int f P 1_A d\lambda = \int f(x) P(x, A) \lambda(dx) \leq c \int P(x, A) \lambda(dx) \leq c\lambda(A).$$

Therefore if  $0 \leq f \in K$  and  $c > 0$  then  $P^*P(\min(f, c)) \leq f$  and  $P^*P(\min(f, c)) \leq c$  hence  $P^*P(f - \min(f, c)) = f - P^*P(\min(f, c)) \geq f - \min(f, c)$ . Inequality is impossible, hence  $f - \min(f, c) \in K$  and therefore  $\min(f, c) \in K$ .

Thus the conditions of [2], Lemma 1, are satisfied and  $K$  is generated by the characteristic functions it contains. Let us conclude the proof by showing that if  $1_A \in K$  then  $P 1_A = P(x, A)$  is a characteristic function:

Put  $f = P 1_A$  then  $0 \leq f \leq 1$  and  $P^*f = 1_A$ .

It is to be

$$1_A(x) = \begin{cases} 1 & \text{for } x \in A, \\ 0 & \text{for } x \in X \setminus A. \end{cases}$$

Now if  $B_\varepsilon = \{x: f(x) > \varepsilon\}$  and  $C$  is a set of finite  $\lambda$  measure disjoint to  $A$  then  $0 \leq \int P^* 1_{B_\varepsilon} \cdot 1_C d\lambda \leq \frac{1}{\varepsilon} \int P^* f \cdot 1_C d\lambda = 0$  or  $P^* 1_{B_\varepsilon} \leq 1_A = P^* f$ . Thus

$$0 \leq \int P^*(1_{B_\varepsilon} - f) \cdot 1_A d\lambda = \int (1_{B_\varepsilon} - f) \cdot f d\lambda \rightarrow \int (1_B - f) \cdot f d\lambda$$

where  $B = \{x: f(x) > 0\}$ . Now  $1_B \geq f$  hence  $f = 1_B$ .

**Lemma 2.** *Let  $P(x, A)$  satisfy the conditions of Lemma 1. The corresponding operator  $P$  on  $L_2$  satisfies (W') provided:*

$$(2.3) \quad P(\cdot, A) = 1_B \text{ and } \lambda(A) < \infty \text{ implies } 1_B = 1_A.$$

**Proof.** Condition 2.3 implies that every characteristic function in  $K$  is left invariant by  $P$  thus  $K$  itself is invariant under  $P$  by Lemma 1.

**Remark.** If  $P(\cdot, A) = 1_B$  this means that whenever there is a positive probability to move from  $x$  to  $A$  then the process moves  $x$  to  $A$  surely. Let us call such a set  $A$  a "trap set". Thus condition (2.3) can be rephrased:

*Every "trap set" captures only its own members.*

**Theorem.** *Let  $r(n)$  assume the values  $1, \dots, N$  infinitely often. Let  $\mu$  be a probability measure which is absolutely continuous with respect to  $\lambda$  and put  $d\mu = f d\lambda$ ,  $0 \leq f \in L_1(X, \Sigma, \lambda)$ . If  $P_j(x, A)$  ( $1 \leq j \leq N$ ) satisfy (2.3) then for every set  $A$  with  $\lambda(A) < \infty$*

$$\lim_{n \rightarrow \infty} (\mu P_{r(0)} \dots P_{r(n)})(A) = \mu_0(A)$$

where  $d\mu_0 = f_0 d\lambda$  and  $f_0$  is the conditional expectation of  $f$  on the field  $\Sigma' = \bigcap_{j=1}^N \{B: \lambda(B) < \infty, P_j(\cdot, B) = 1_B\}$ .

**Proof.** With no loss of generality we may assume that  $f \in L_2(X, \Sigma)$ . It is well known that the conditional expectation  $f_0$  of  $f$  on the field  $\Sigma'$  is equal to the orthogonal projection  $P'f$  of  $f$  on the subspace  $L'$  of  $L_2(X, \Sigma, \lambda)$ , spanned by  $\Sigma'$ .

By the Amemiya—Ando theorem  $P_{r(0)} \dots P_{r(n)}$  and therefore  $P_{r(n)}^* \dots P_{r(0)}^*$  converge weakly to the orthogonal projection  $P'$  on  $L'$ . Thus

$$\begin{aligned} (\mu P_{r(0)} \dots P_{r(n)})A &= \int (P_{r(0)} \dots P_{r(n)} 1_A) d\mu = \\ &= \int (P_{r(0)} \dots P_{r(n)} 1_A) f d\lambda = \int 1_A P_{r(n)}^* \dots P_{r(0)}^* f d\lambda \end{aligned}$$

tends to  $\int_A P' f d\lambda = \int_A f_0 d\lambda = \mu_0(A)$ .

Let us consider the case where the only "trap sets", for the subtransition functions  $P_j(x, A)$ , are the trivial sets. Then if  $\lambda$  is not finite,  $K = \{0\}$  and  $\lim (\mu P_{r(1)} \dots P_{r(n)})(A) = 0$  whenever  $\lambda(A) < \infty$ . On the other hand if  $\lambda(X) < \infty$  then  $K = \{\text{constant functions}\}$  and  $\lim_{n \rightarrow \infty} (\mu P_{r(1)} \dots P_{r(n)})(A) = \lambda(A) \mu(X) / \lambda(X)$ .

### References

- [1] I. AMEMIYA and T. ANDO, Convergence of random products of contractions in Hilbert space, *Acta Sci. Math.*, **26** (1965), 239–244.
- [2] S. R. FOGUEL, On order preserving contractions, *Israel J. Math.*, **1** (1963), 54–59.

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